dV/dt Accelerating the Rate of Progress towards Extreme Scale Collaborative Science

Miron Livny (UW)
Ewa Deelman, Gideon Juve, Rafael Ferreira da Silva (USC)
Ben Tovar, Casey Robinson, Douglas Thain (ND)
Frank Wuerthwein (UCSD)
Bill Allcock (ANL)

Funded by DOE



Thesis

- Researchers band together into dynamic collaborations and employ a number of applications, software tools, data sources, and instruments
- They have access to a growing variety of processing, storage and networking resources
- Goal: "make it easier for scientists to conduct large-scale computational tasks that use the power of computing resources they do not own to process data they did not collect with applications they did not develop"





Challenges today

- Estimate the application resource needs
- Finding the appropriate computing resources
- Acquiring those resources
- Deploying the applications and data on the resources
- Managing applications and resources during run
- Make sure the application actually finishes successfully!

 Approach: Develop a framework that encompass the five phases of collaborative computing—estimate, find, acquire, deploy, and use



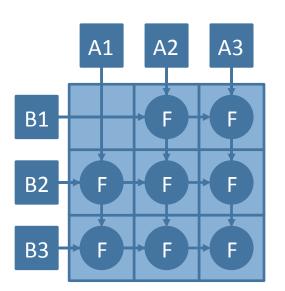


Application Characterization

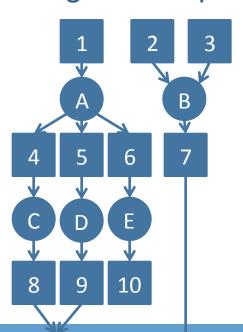
Concurrent Workloads

Static Workloads

Regular Graphs



Irregular Graphs



Dynamic Workloads

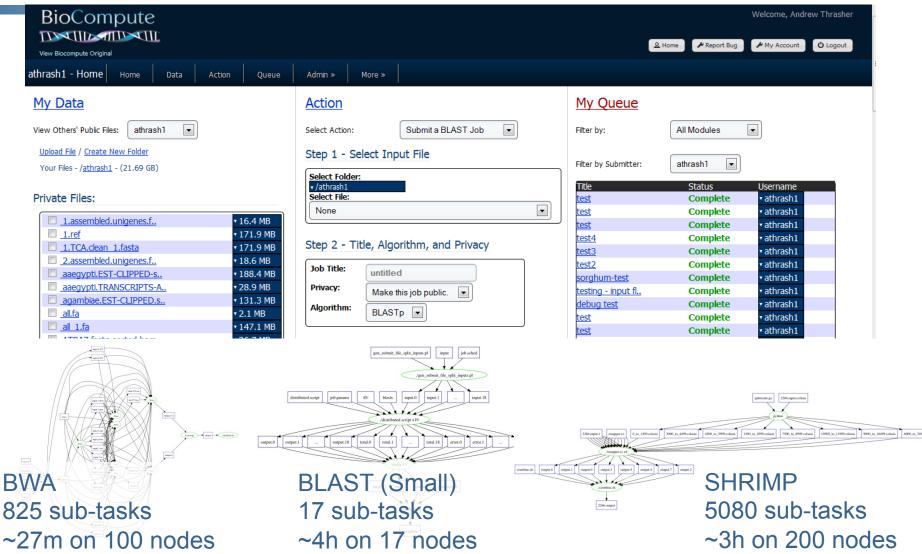
```
while( more work to
do) {
    foreach work unit {
        t = create_task();
        submit_task(t);
    }

    t = wait_for_task();
    process_result(t);
}
```





Portal Generated Workflows using Makeflow



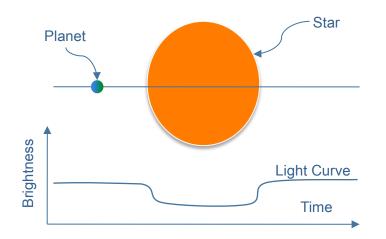


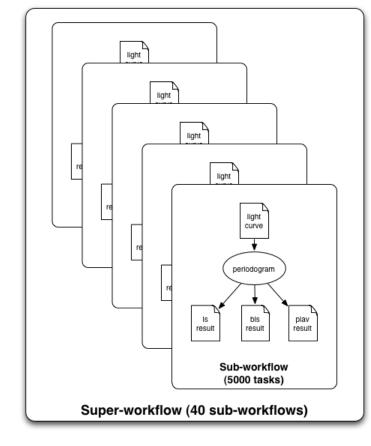


Periodograms: generate an atlas of extra-solar planets

- Find extra-solar planets by
 - Wobbles in radial velocity of star, or
 - Dips in star's intensity

210k light-curves released in July 2010Apply 3 algorithms to each curve3 different parameter sets



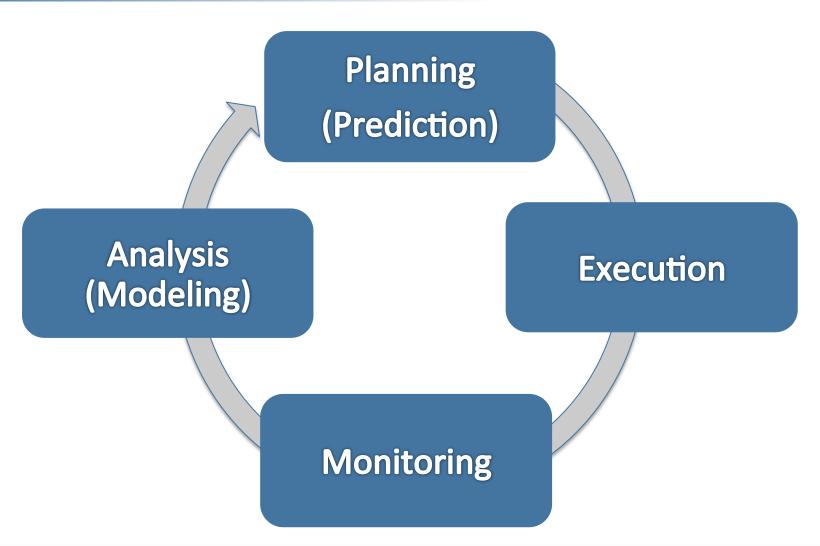


- 210K input, 630K output files
- 1 super-workflow
- 40 sub-workflows
- ~5,000 tasks per sub-workflow
- 210K tasks total





Characterizing Application Resource Needs





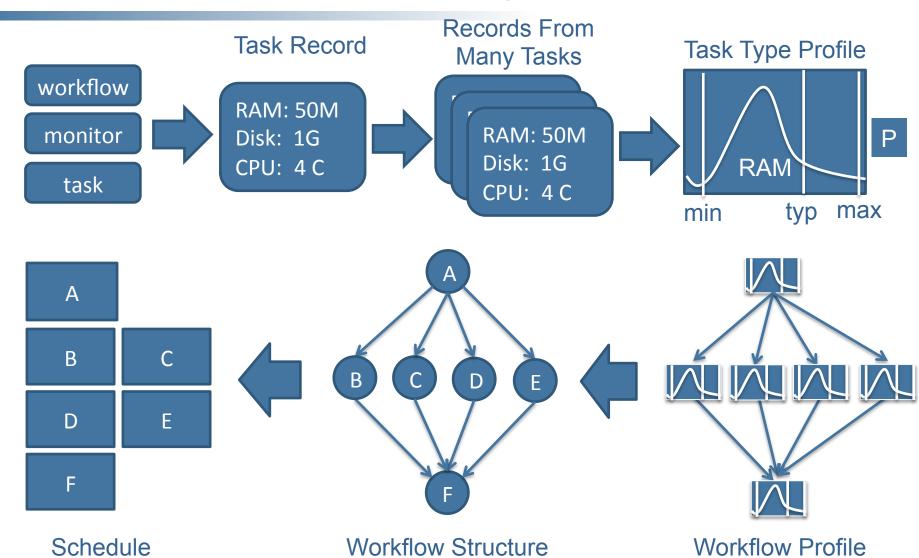


Task Characterization/Execution

- Understand the resource needs of a task
- Establish expected values and limits for task resource consumption
- Launch tasks on the correct resources
- Monitor task execution and resource consumption, interrupt tasks that reach limits
- Possibly re-launch task on different resources



Data Collection and Modeling







Resource Usage Monitoring





Resource Monitoring

Measure Resource Usage

- Runtime (wall time of process)
- CPU usage (FLOPs, utime, stime)
- Memory usage (peak resident set size, peak VM size)
- I/O (data read/written, number of reads/writes)
- Disk (size of files accessed/created)

Impose Limits

- Use models to predict usage
- Use predictions to set limits
- Detect violations of limits to prevent problems at runtime





Monitoring Accuracy with Synthetic Benchmarks

	D 1:										
	Baseline	Polling			fork/exit			z/exit	syscall		
			•. \	LD_PRELOAD			ptrace		ptrace		
		(resource	ce_monitor)	`	esource_monitor)			start)	(kickstart)		
Instr.				(a) CPU							
10^{6}	$0.32 \mathrm{\ s}$	+0.04	(12.50%)	+0.02	(4.91%)	(0.00	(0.00%)	0.00	(0.00%)	
10^{7}	$2.93 \mathrm{\ s}$	+0.06	(2.12%)	+0.04	(1.20%)	(0.00	(0.00%)	+0.01	(0.14%)	
10^{8}	28.20 s	+0.17	(0.60%)	+0.09	(0.31%)	+(0.03	(0.10%)	+0.04	(0.14%)	
10^{9}	279.53 s	+1.29	(0.46%)	+1.32	(0.47%)	+(0.20	(0.07%)	+0.41	(0.15%)	
Memory			(b) M	emory:	resident siz	ze					
1GB	1GB		-13.96%		+0.08%			+0.03%		+0.03%	
2GB	2GB		-17.63%		+0.03%			+0.02%		+0.02%	
4GB	4GB		-2.25%		+0.02%			0.00%	0.00%		
8GB	8GB		-1.89%		+0.01%			0.00%	0.00%		
16GB	16GB		-1.99%		+0.01%			0.00%		0.00%	
File size			(c) I/O:	bytes re	ead, 4KB bi	<mark>u</mark> ffe	er				
1MB	1MB		-13.64%		0.00%			0.00%		0.00%	
100MB	100MB		-9.07%		0.00%			0.00%		0.00%	
1GB	1GB		-5.84%		0.00%			0.00%		0.00%	
10GB	10GB		-2.13%		0.00%			0.00%		0.00%	
Buffer size			(d) I/O	: bytes	read, 1GB	file	;				
4KB	1GB		-5.84%		0.00%			0.00%		0.00%	
8KB	1GB		-0.82%		0.00%			0.00%		0.00%	
16KB	1GB		-15.41%		0.00%			0.00%		0.00%	
32KB	1GB		-18.41%		0.00%			0.00%		0.00%	





Monitoring Overhead

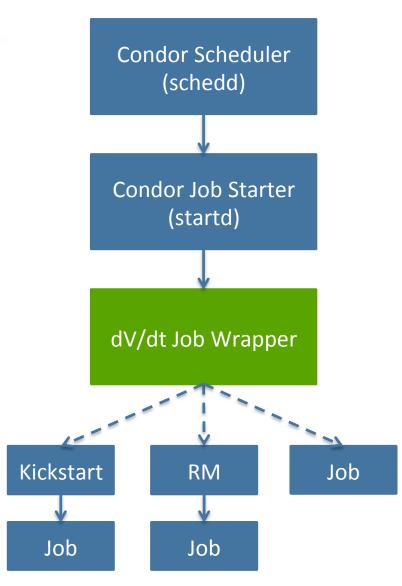
	Baseline	Po	olling		ork/exit	//	rk/exit	syscall			
				LD_F	PRELOAD	ptrace		ptrace			
		(resource	ce_monitor)	(resour	ce_monitor)	(ki	ckstart)	(kickstart)			
Instr.			(a) CPI overhead								
10^{6}	0.32 s	+0.22	(68.75%)	+0.25	(78.13%)	+0.18	(56.25%)	+0.13	(40.63%)		
10^{7}	2.93 s	+0.28	(9.56%)	+2.42	(82.59%)	+0.14	(4.78%)	+0.14	(4.78%)		
10^{8}	28.20 s	+0.17	(0.60%)	+0.22	(0.78%)	+0.10	(0.35%)	+0.12	(0.43%)		
10^{9}	279.53 s	+0.28	(0.10%)	+0.78	(0.28%)	+0.07	(0.03%)	+0.61	(0.22%)		
Resident size					(b) Memo	ory over	head				
1GB	$3.57 \mathrm{\ s}$	+0.17	(4.76%)	+0.26	(7.28%)	+0.06	(1.68%)	+0.07	(1.96%)		
2GB	6.19 s	+0.10	(1.62%)	+0.14	(2.26%)	+0.09	(1.45%)	+0.06	(0.97%)		
4GB	12.64 s	+0.50	(3.96%)	+0.86	(6.80%)	+0.24	(1.90%)	+0.43	(3.40%)		
8GB	25.06 s	+0.51	(2.04%)	+1.88	(7.50%)	+0.87	(3.47%)	+0.96	(3.83%)		
16GB	52.81 s	+1.11	(2.10%)	+4.69	(8.88%)	+1.38	(2.61%)	+2.25	(4.26%)		
File size				`	c) I/O overh	ead, 4K					
1MB	$0.01~\mathrm{s}$	+0.17	(1700%)	+0.24	(2400.00%)	+0.13	(1300.00%)	+0.14	(1400.00%)		
100MB	1.53 s	+0.09	(5.88%)	+0.10	(6.54%)	+0.09	(5.88%)	+1.82	(118.95%)		
1GB	16.02 s	+0.04	(0.25%)	+0.38	(2.37%)	+0.36	(2.25%)	+15.98	(99.75%)		
10GB	153.98 s	+0.54	(0.35%)	+0.64	(0.42%)	+0.58	(0.38%)	+143.95	(93.49%)		
Buffer size			(d) I/O overhead, 1GB file								
4KB	16.02 s	+0.04	(0.25%)	+0.38	(2.37%)	+0.36	(2.25%)	+15.98	(99.75%)		
8KB	9.14 s	+0.20	(2.19%)	+0.38	(4.16%)	+0.24	(2.63%)	+8.72	(95.40%)		
16KB	6.40 s	+0.23	(3.59%)	+0.34	(5.31%)	+0.30	(4.69%)	+4.13	(64.53%)		
32KB	4.37 s	+0.18	(4.12%)	+0.43	(9.84%)	+0.60	(13.73%)	+2.11	(48.28%)		





Condor Job Wrapper

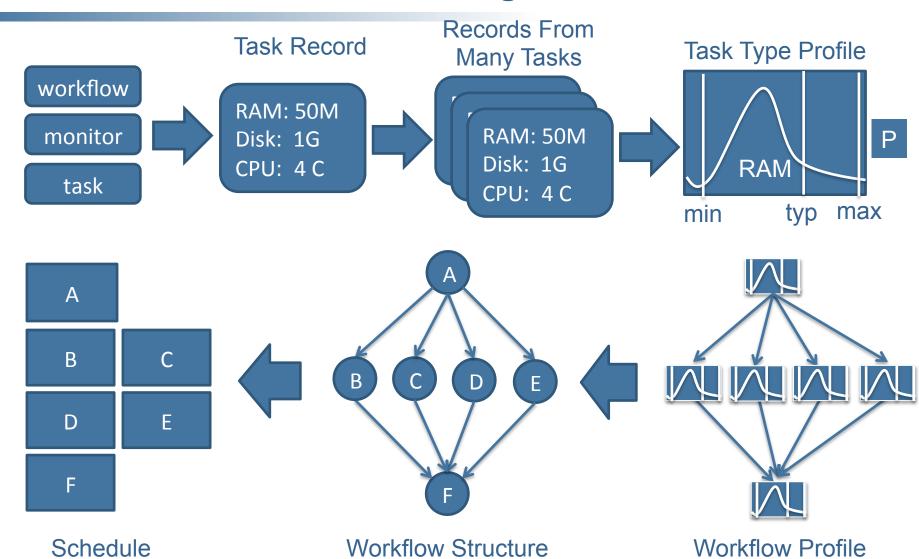
- Selectively wraps Condor jobs with monitoring tools
 - Uses USER_JOB_WRAPPER functionality of Condor
 - Does not wrap jobs that have failed
 - Selectively monitors based on user, executable, etc.
 - Selectively monitors a given percentage of jobs (e.g. 50% of jobs)
 - Detects monitor errors and restarts job without wrapper
- Allows us to easily deploy monitoring tools on production Condor pools







Data Collection and Modeling







Resource Monitoring Archive

- Stores monitoring records
- Provides a query interface for analyzing data

resource	wall time	cpu time	resident memory		
	21490	21022	61615		
$\operatorname{histogram}$	122 s 321s 777 s	121 s 319 s 684 s	208 MB 817 MB		
mean	410.55 s	$406.17 \; \mathrm{s}$	682.62 MB		
std. dev.	79.16	73.86	208.83		
skewness	0.42	0.17	-1.11		
kurtosis	0.26	-0.10	10.96		





Resource Usage Limits

memory: 4000000 num_proc: 300

cpu_time: 40000000

disk: 100000000000

global: limits file

#: num_proc: 250, fs_nodes: 1 sqrts: msqrt

./msqrt 200

local: per task rule

Limits specification

command: ./msqrt 200

start: 1361995712680901 Wed Feb 27 15:08:32 2013 end: 1361995725794759 Wed Feb 27 15:08:45 2013

exit-type: signal 2 Termination

exit-status: 1

monitor-watch-end: cpu_time 42740000 > 40000000

max

processes: 201

cpu_time: 42740000 memory: 410295

io-chars: 67467528

vnodes: 22

bytes: 94187

fs_nodes:

Record with alarm





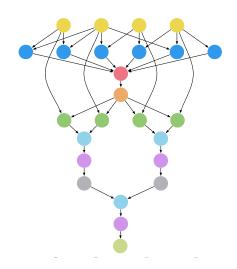
Resource Usage Modeling





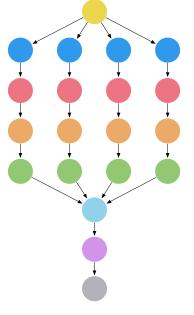
Workflow Execution Profiling

- Workflows were executed using <u>Pegasus WMS</u> and profiled
 - Monitors and records fine-grained data
 - E.g. process I/O, runtime, memory usage, CPU utilization
- 3 runs of each workflow with different datasets





Periodogram Workflow



Epigenomics Workflow







Execution Profile: Montage Workflow

Task estimation could be based on mean values

		rask estimation codia be based on mean values									
						K		K			
Task	Count		ntime	I/O Read			I/O Write			Mexnory Peak	
Idak	Count	Mean (s)	Std. Dev.	Mean (MB)	Std. Dev.	Mean	(MB)	Std. Dev.	Me	an (MB) S	td. Dev.
mProjectPP	7965	2.59	0.69	4.24	0.19		16.20	0.80		9.96	0.40
mDiffFit	23733	1.25	0.92	24.08	5.76		1.35	1.11		5.32	0.90
mConcatFit	3	122.04	5.27	2.70	0.01		3.15	0.01		7.26	0.01
mBgModel	3	2008.08	88.50	4.14	0.04		0.27	0.00		14.41	0.01
mBackground	7965	2.14	1.68	13.67	6.78		13.05	6.44		11.75	5.78
mImgtbl	51	4.65	2.04	22.64	4.61		0.25	0.05		6.37	0.13
\mathbf{mAdd}	51	47.69	14.03	2191.76	560.39	15	574.22	383.86		21.66	3.40
mShrink	48	11.53	2.25	835.57	0.31		1.00	0.00		3.05	0.01
mJPEG	3	1.03	0.07	46.18	0.02		0.78	0.00		2,86	0.01
uses Kickstart profiling tool											

Task estimation based on average may lead to significant estimation errors

16-core cluster

- 5 Dual core MP Opteron™ Processor 250 2.4GHz / 8GB RAM
- 3 Dual core MD AMD OpteronTM Processor 275 2.2 GHz / 8GB RAM





Automatic Workflow Characterization

- Characterize tasks based on their estimation capability
 - Runtime, I/O write, memory peak → estimated from I/O read

- Use correlation statistics to identify statistical relationships between parameters
 - High correlation values yield <u>accurate estimations</u>, <u>Estimation based</u>
 on the ratio: <u>parameter/input data size</u>

Task	Run	time	I/O	Write	Memor	y Peak	•
Idak	ρ	σ	ρ	σ	ρ	σ	Constant values
fastqSplit	0.98	9.00	1.00	297.15	0.00	0.01	7 Solitain Values
filterContams	-0.03	0.27	0.99	1.46	0.00	0.01	
sol2sanger	0.21	0.41	0.90	1.49	0.00	0.01	
fast2bfq	0.18	0.27	0.56	0.87	0.00	0.01	Correlated if
map	0.02	18.96	0.06	0.70	0.01	1.43	$\rho > 0.8$
mapMerge	0.98	13.33	0.99	189.81	-0.36	2.15	P
pileup	0.99	4.73	0.17	249.78	0.87	25.70	

Epigenomics workflow





Task Estimation Process

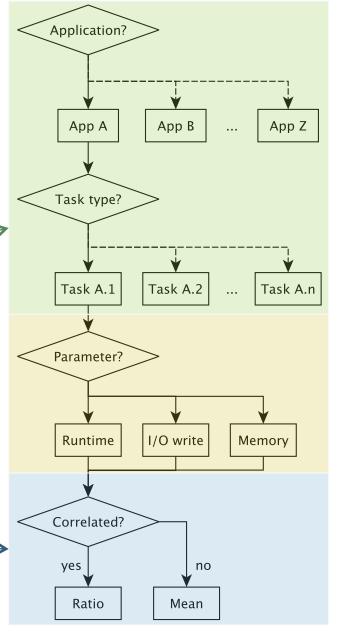
- Based on Regression Trees
 - Built offline from historical data analyses

Tasks are classified by application, then task type

Estimation of runtime, I/O write, or memory peak

If strongly correlated to the input data:

- Estimation based on the ratio parameter/input data size
- Otherwise, estimation based on the <u>mean</u>

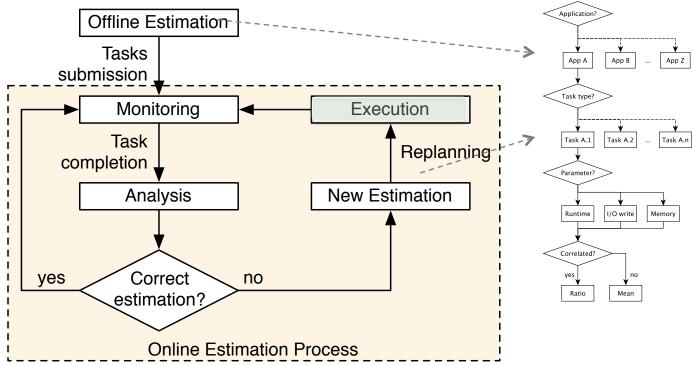






Online Estimation Process

- Based on the MAPE-K loop
 - Task executions are <u>constantly monitored</u>
 - Estimated values are updated, and a <u>new prediction</u> is done







Experiment: Use Estimations Online, while the workflow is executing

- Trace analysis of 3 workflow applications
 - Montage
 - Epigenomics
 - Periodogram

- Leave-one-out cross-validation
 - Evaluate the accuracy of our online estimation process
 - 3 different workflow execution traces for each workflow
- Simulator
 - Replays workflow executions





Results: Average Estimation Errors - Montage

Task	Estimation	Runtime Avg. Error (%)	I/O Write Avg.Error (%)	Memory Avg.Error (%)
mProjectPP	Offline	18.95	1.63	2.80
	Online	18.95	1.63	2.80
mDiffFit	Offline	191.02	159.46	91.07
	Online	46.52	69.14	73.72
mConcatFit	Offline	4.38	0.00	7.62
	Online	4.03	0.00	6.22
mBgModel	Offline	23.83	0.00	22.08
	Online	1.17	0.00	3.43
mBackground	Offline	65.13	102.80	104.62
	Online	44.90	1.23	1.84
mImgtbl	Offline	61.27	127.29	126.58
	Online	29.15	5.53	8.35
mAdd	Offline	9.67	113.14	110.20
	Online	9.31	3.43	9.06
mShrink	Offline	13.72	0.34	0.00
	Online	7.61	0.33	0.00
mJPEG	Offline	1.61	0.00	19.09
	Online	1.37	0.00	11.40

Poor output data estimations leads to a chain of estimation errors in scientific workflows

Offline Process

Avg. Runtime Error: 43% Avg. I/O Write Error: 56% Avg. Memory Error: 53%

Online Process

Avg. Runtime Error: 18% Avg. I/O Write Error: 9% Avg. Memory Error: 13%

Online strategy counterbalances the propagation of estimation errors





Conclusions

A planning framework that:

- Starts with an unknown application
- Characterizes it, models it, and manages execution dynamically

Future:

- Experiments at scale on Condor pool at UW and OSG resources (model heterogeneous resources)
- Integrate resource provisioning into planning
- Experiment with predictions and resource provisioning
- https://sites.google.com/site/acceleratingexascale/



